

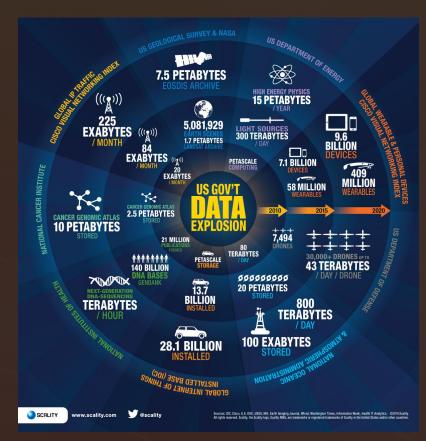






#### **Modern Big Data Applications**

- Explosion of data volume, variety, and velocity
  - Facebook is storing roughly 250 billion images.
  - 80% of all the world's data is unstructured.
  - Facebook users upload more than 900 million photos a day.
  - Square Kilometer Array (SKA) is estimated to reach 10 Pb/s.



Generates tremendous stress on storage sub-system



# Data Compression is popularly used to ease this stress



#### **Data Compression**

- Categories of compression techniques
  - lossy and lossless algorithms.
- The lossless algorithms are standard in scientific and cloud applications.
- Popular examples of lossless algorithms
  - General Purpose: Bzip, Zlib, 7z, etc.
  - Specialized: Snappy, SPDP, LZO, etc.





### **Challenges in Data Compression**

- <u>Data-dependency:</u> Each Compression library is specialized for a certain input (i.e., data-type and data-format)
- <u>Library-choice</u>: Choice of library is complex as different situations might demand different compression needs.
- API diversity: Each library has its own definition of Interface.

These challenges highlight there is no "one compression for all".

Can we do something better?





# **Outline**

Ares

- Approach
- <u>Design</u>
- Results
- Conclusion



# **Approach: Overview**

Benchmark

**Build Ares** 



Analyze





**Conclusion** 

#### **Approach: Benchmark**

- <u>Library Corpus:</u> bzip2, zlib, huffman, brotli, bsc, lzma, lz4, lzo, pithy, snappy, and quicklz.
- <u>Data-Types:</u> characters, integers along with their modifiers (short, long, signed, unsigned), sorted integers, floating point, and double floating points.
- Data-Formats: binary, HDF5, csv, json, xml, and Avro, Parquet.
- Metrics: Compression/Decompression Speed & Compression Ratio

These total to over 1000 test cases



**Conclusion** 

#### **Approach: Analyze**

- Workload Priority: defines different requirement that a workload prioritizes
- Score Formulation:

Binary Format								
cs	DS	CR	Workload	Char	Integer	Sorted int	Float	Double
1	0	0	Asynchronous communication	Iz4	lz4	lz4	quicklz	lz4
0	1	0	Multicast in Network	Iz4	lz4	pithy	pithy	brotli
0	0	1	Archival Store	bsc	Izma	bsc	Izma	bsc
0.5	0.5	0	Synchronous Communication	Iz4	lz4	pithy	pithy	lz4
0	0.5	0.5	Dequeue Operation	lz4	lz4	lz4	quicklz	pithy
0.5	0	0	Queue Operation	Iz4	lz4	lz4	pithy	lz4
0.3	0.3	0.3	Mixed workload	lz4	lz4	pithy	pithy	pithy





#### **Approach: Build Ares**

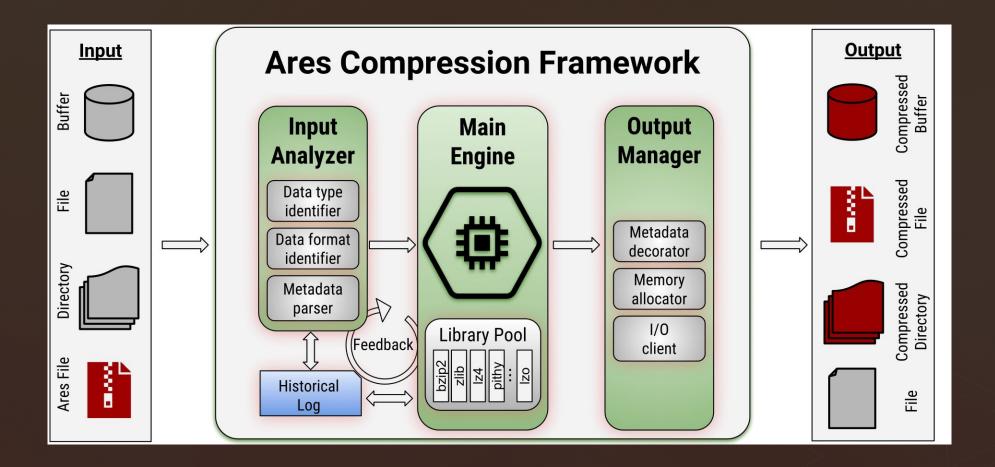
#### Goals

- The framework should be able to learn and adjust itself to the input data compression characteristics.
- The framework should be able to reconfigure itself, dynamically, to various compression needs of an application.
- The framework should be able to unify all interfaces of the compression libraries it contains.

# **Ares Design**



# **Design: Overview**



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Approach

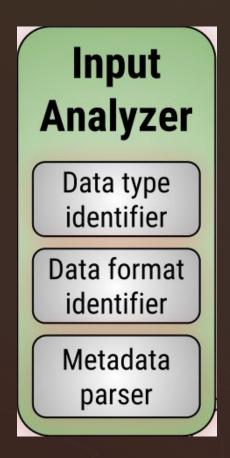
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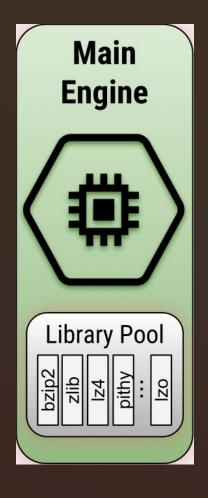


- Infers data type and format
- Uses a hybrid approach
  - static analysis and a dynamic feedback mechanism
- <u>Data-format:</u> mime-type, extensions, and metadata-rich.
- <u>Data-type:</u> decoding techniques, type-inference and metadata-rich.



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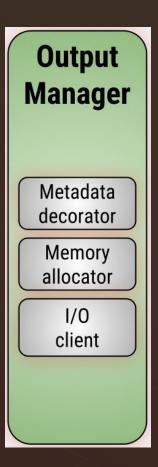


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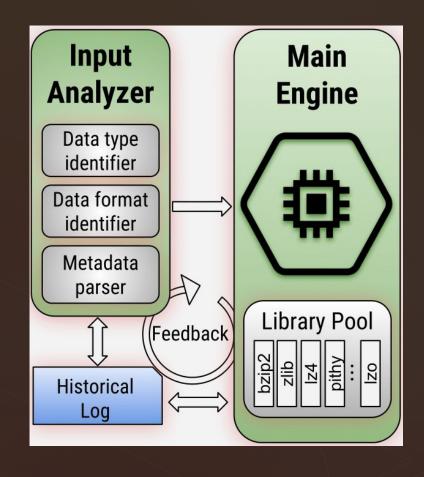
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- Decorates the compressed data with headers, regarding the compression library used
  - 8 bytes header per data-type
- Checks the correctness of the format using parity checking
- Performs final I/O of the compressed/uncompressed data



- The engine updates the log with actual performance results
- Analyzer processes the log to identify the difference between expected and actual measurements.
- This makes the analyzer improve it predictions over time.



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#### **Evaluation**



#### **Evaluation: Testbed**

#### Machine Configuration (Per Node)

- dual Intel(R) Xeon(R)
  - CPU E5-2670 v3
  - 2.30GHz
  - 48 cores
- 128 GB RAM
- 10 Gbit Ethernet,
- 200 GB HDD

#### Deployment

- Scientific Setup:
  - 32 client nodes
  - 8 PFS nodes
- Cloud Setup:
  - 40-node Hadoop cluster
  - 1 Namenode

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#### **Evaluation: Goals**

- Overheads and Resource Utilization
  - Ares's analysis overheads + CPU + Memory utilization
- Compression/Decompression Intelligence
  - Data type and format aware data compression
- Compression/Decompression Adaptiveness
  - Workflow-specific data compression
- Compression/Decompression Flexibility
  - Ares for various real applications

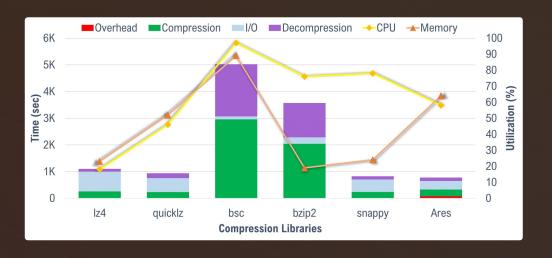


# **Evaluation: Overheads**

#### Description:

SCALABLE COMPUTING

- 64GB HDF5 input with four datasets: characters, integers, sorted integers, and doubles.
- Workflow: read input data -> compress data -> write compressed data -> read compressed data -> decompress the data.
- Metrics overall time and utilization



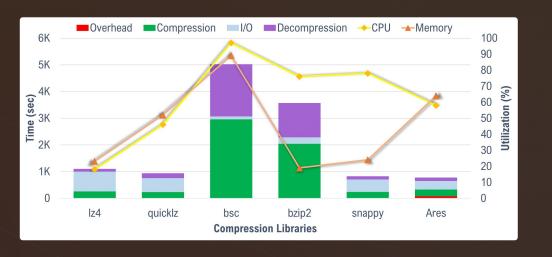




#### **Evaluation: Overheads**

#### Observations

- libraries demonstrates different overheads
- Ares balances the tradeoff between CT/DT and CR by analyzing the input data with a 10% overhead.
- Ares performs better as it uses a collection of libraries where they have strength.





Conclusion

# **Evaluation: Intelligence**

- Description:
  - Different Data-Types
  - 64 GB of buffer input
  - Configurations of this buffer:
    - Characters, integers, floats, doubles, and a mixed case
- We measure the CT, DT and CR.





# **Evaluation: Intelligence**

#### Observation:

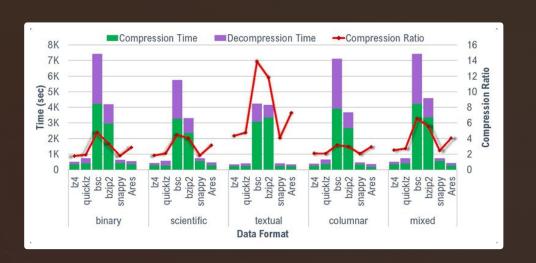
- Different libraries excel in different data types.
- trade-off between CT and CR
- For mixed input each library takes a hit in performance
  - Ares optimizes by using best library for given data-type.





### **Evaluation: Intelligence**

- Description:
  - Different Data-Format
  - 64 files (each 1 GB) in a directory
  - Composition of this folder:
    - POSIX , HDF5, pNetCDF, HTML,
      XML, JSON, Avro, and Parquet
- We measure the CT, DT and CR.

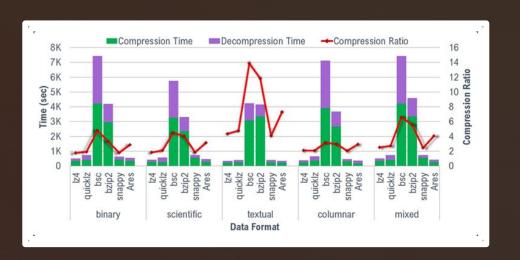




#### **Evaluation: Intelligence**

#### Observation:

- Different libraries excel in different data formats.
- trade-off between CT and CR
- For mixed directory each library takes a hit in performance
  - Ares optimizes by using best library for given data-format.





#### **Evaluation: Adaptiveness**

- Description:
  - Different Workflow Priorities
  - 64 GB of CSV file input
  - Four columns of this file:
    - Index (sorted integer), location (char), population size (integer), income (double)
- We measure the CT, DT and CR.







**Conclusion** 

#### **Evaluation: Adaptiveness**

#### Observation:

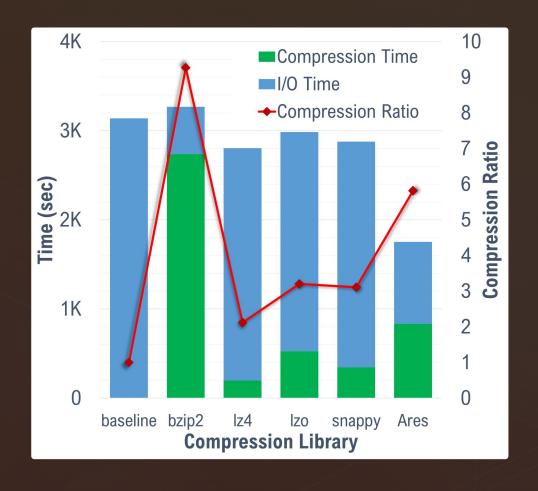
- This multi type data has an effect on every compression library
- Different prioritization of Ares results in difference in performance metrics
- In Balanced mode, Ares is a Jack of all trades.





# **Evaluation: Scientific Application (VPIC)**

- VPIC simulation
  - Each process is producing 1 GB at each time step.
  - The overall data size is 1.5 TB
  - HDF5 file is organized with 7 datasets
    - two datasets of integers, two of floats and three of doubles.
- We show Compression Time,
  Compression Ratio and I/O Time.



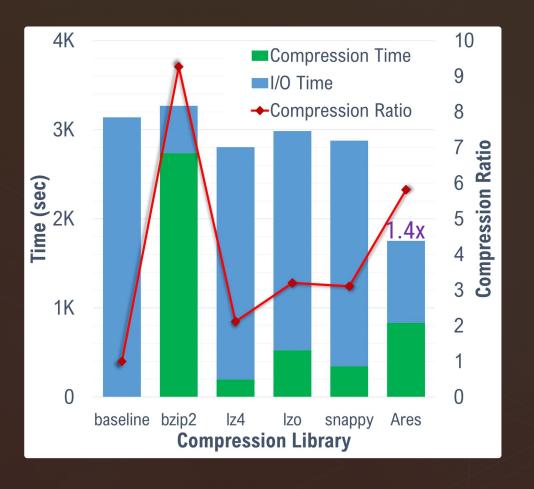
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### **Evaluation: Scientific Application (VPIC)**

#### Observation

- Compression reduces the I/O time.
- Heavy compression is costly, and a balance must be found to be beneficial to the application
- Opposite picture can be seen when using lz4, lzo, and snappy as compression filters.
- Ares prioritizes both CT and CR.



Approach

Design

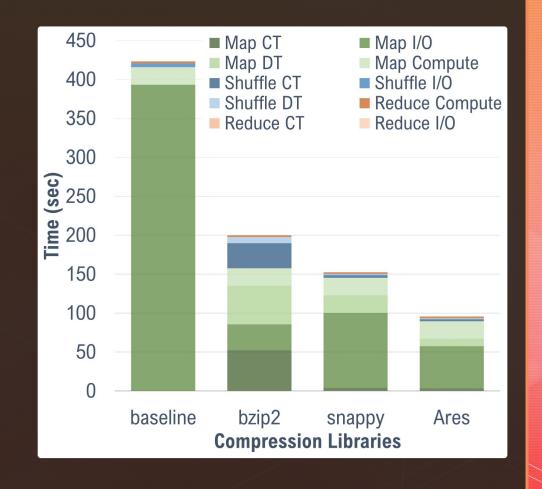
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# **Evaluation: Hadoop Application (Word Count)**

- Map-Reduce implementation of the word-count kernel 32 mappers and 8 reducers
  - 1.5 TB of HTML files (Wikipedia articles)
  - Workflow:
    - MAP: reads its input data and counts individual word occurrences and create intermediate files
      - a compressed input and a high DS.
    - SHUFFLE: all intermediate files are sorted
      - quick compression to minimize I/O traffic.
    - REDUCE: merge the final count across all intermediate files and write the final word count back to a file in HDFS



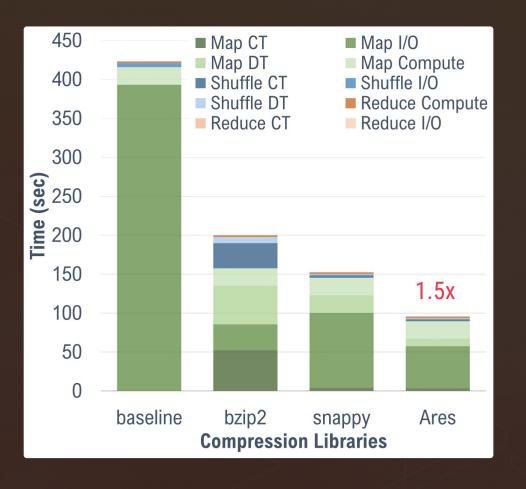
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# **Evaluation: Hadoop Application (Word Count)**

- Observations:
  - compression on the input data reduces the I/O time in map phase.
  - tradeoff CT/DT and CR
  - Ares achieving the best overall performance
  - highlights the importance of striking a balance of compression speed and ratio.
    - Compression libraries do not offer dynamic adaptiveness based on the workload type.



Results

Approach Design

Conclusion



#### **Conclusions**

- We investigated how different data-types, data-format, and workload characteristics affect the choice of the "ideal" compression library.
- We have developed Ares, a dynamic, adaptive, and flexible compression framework, that can transparently meet various compression needs of big data applications.
- Under real world applications Ares performed 2-6x faster than competitive solutions with a 10% analysis cost.

Results



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Q & A





